

Development of Tailored Fiber Placement, Multifunctional, High-Performance Composite Material Systems for High Volume Manufacture of Structural Battery Enclosure

Venkat Aitharaju
General Motors
2021 Annual Merit Review
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Project ID: MAT198

## Overview

#### **Barriers and Technical Targets**

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#### **Timeline**

• Project Start Date: April 1, 2021\*

• Project End Date: June 30, 2024

• Percent Complete: 2%

### **Budget**

• Total project funding

– Total: \$10,330,000

– DOE Share: \$7,500,000

Contractor Share: \$2,830,000

• Funding for FY21:

- Total: \$1,149,000

– DOE share: \$834,174

Contractor share: \$314,826

#### **Barriers addressed\***

- A. Material systems development: Structural composite material system having multi-functional capabilities such as hybrid fibers (ex. carbon and glass), self-health monitoring, fire-retardance, and electro-magnetic compatibility (EMC) to make a positive business case (cost increase per pound saved is less than \$5)
- **B.** Predictive technology development: Modeling tools to predict the performance of manufacturing process and structural design within 15% of experimental results. AI/ML technology development for process monitoring to save costs of inspection and scrap during manufacturing.
- *C. Demonstration:* Using the developed material systems, design, build, and test a structural composite battery enclosure, and compare the weight, performance metrics with that of a baseline metallic assembly.

\*2017 U.S. DRIVE Roadmap Report, section 4

**General Motors** 

**Participants** 

**Coats** 

**Columbia University** 

**Continental Structural Plastics (CSP)** 

ESI Group, NA

**Michigan State University** 

**University of Southern California** 

\*requested DOE a new timeline for the project due to late start

## Relevance

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#### **High Performance Multi-functional Composite Material Systems**

- Hybrid carbon and glass fibers in various architectures for cost-effective business case (≥ 25 Msi, strain ≥ 1%)
- Self-health monitoring electronic circuits embedded in the composite for value proposition
- Integrated fire-retardant material systems for use in the future design of components for the battery enclosures
- Electro-magnetic compatibility for value proposition and use for the future design of battery enclosures
- High-pressure resin transfer molding (HP-RTM) for volume manufacturing needed for automotive industry
- Predictive computational tools for virtual design and eliminating cost of trial-and-error iterations
- AI/ML technologies for process monitoring and component design
- Reduce the lead time and costs to accelerate the implementation of structural automotive composites.
- Enable the usage of composites for significant light-weighting of automobiles and thus improve fuel economy/range, and lower emissions (reduce greenhouse gas emissions).

#### **Cost Barrier**

• Will demonstrate the ability to manufacture the automotive composite assembly at no more than **\$5 per pound saved** (2010 dollars).

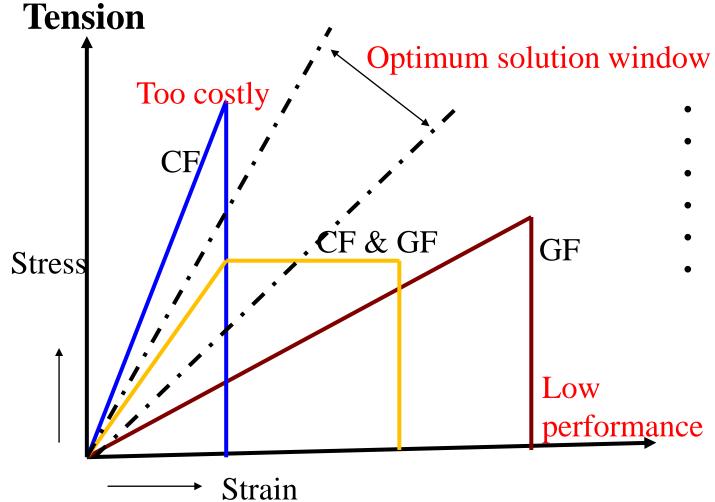
#### **Performance Barrier**

• Will demonstrate the viability of composite materials to meet vehicle performance requirements while reducing vehicle assembly weight **by 25%** compared to a current steel structural battery enclosure. The cycle time to manufacture the composite panels need to be less than 3 minutes.

## Relevance



## Hybrid Fiber Composites



- CF is >10 times costly than GF
- Strength ratio of GF\*/CF\$ = 3448/4157 = 0.83
- Stiffness ratio of  $GF*/CF^{\$} = 72.4/242 = 0.30$
- Load case interest strength based
- Domain of interest beyond the initial failure
- A flexure, crashworthiness performance case will be interesting with more glass fiber content as glass fibers when compared to carbon fibers has good strength properties compared to stiffness.

## Relevance

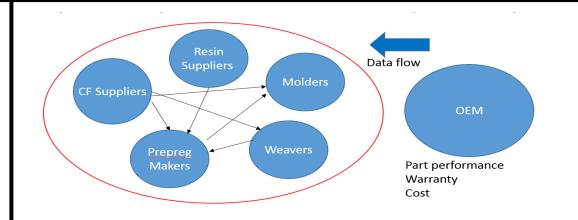
#### **Steps in implementing CF in Automobiles**

## Workflow between OEM and Suppliers Current

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#### **Current**

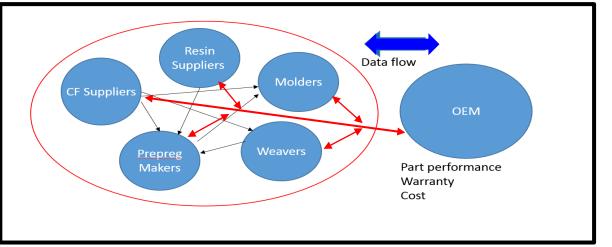
- Design.
- Selection of manufacturing process.
- Manufacturing feasibility.
- Prototype build and learn.
- Modify design and manufacturing process, if needed.
- Improve prototype build and make parts.
- Extrapolate to high volume manufacturing.
- Build the part in high volume, iterate to get good quality.
- Evaluate the performance and compare with requirements.
- If failure occurs, redesign the part.



#### **Future**

- Design.
- Virtual manufacturing simulation and improve the design for optimizing the cost.
- Include manufacturing outcome in performance simulation and further optimize the design to meet the requirements.
- Build tools, manufacture parts and check the performance

#### **Future**



## Milestones



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Task	Details
Task 1.1	Determine functional requirements of baseline assembly
Task 1.2.1	Evaluate and develop an optimized fiber material systems
Task 1.2.2	Evaluate and develop required manufacturing model
Task 1.2.3	Evaluate and develop required structural model
Task 1.2.4	Evaluate and develop required self-health monitoring technologies
Task 1.2.5	Evaluate and develop AI/ML technologies for monitoring the manufacturing process
Task 1.2.6	Evaluate and develop cost models

Task	Details
Task 1.3	Initial structural design of battery enclosure
Task 1.4	Initial manufacturing design of battery enclosure
Task 1.5	Initial self-health monitoring technology implementation
Task 1.6	Initial AI/ML technology implementation

## Approach/Strategy



- Hybrid fiber material system to lower the material costs utilize the ductility improvement cited in the recent literature
- Optimum hybrid fiber ratio (carbon and glass) to maximize the performance for a given cost
- Engineer the microstructure (spacing of fiber bundles, stitch density) of the fiber preform to optimize the performance such as draping, injection (enhanced permeability)
- Multi-functional material systems including fire-retardance and EMC performance
- Develop the technology for high-pressure resin transfer molding (HP-RTM) process
- Develop the technology for self-health monitoring
- Develop the AI/ML technology for process monitoring
- Predictive modeling tools for the developed material systems for both manufacturing and structural performance
- Demonstrate the technology development by design, building, testing and comparing the performance metrics (weight, cost, performance) with that of a baseline metallic assembly.



## FY 21 Accomplishments

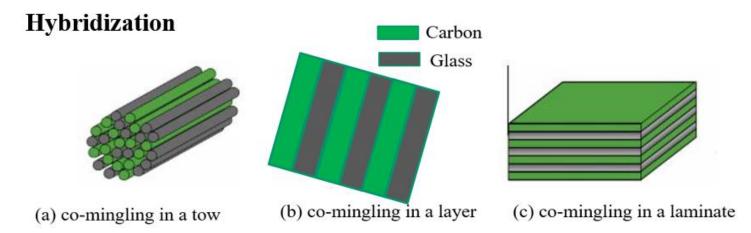
- Determined the functional requirements of the baseline metallic assembly (Task 1.1).
- Finalized the design of experiments test matrix (full factorial) for determining the optimum hybrid material system for performance, draping, injection and structural performance. The same data will be used to develop predictive models.
- Finalized test matrix for the characterization of conductive wiring for developing the self-health monitoring technology.
- Developed a demonstration problem for the AI/ML process monitoring technology.



#### Baseline Model, functional requirements of the battery enclosure

- Steel based design for the battery enclosures
- Performance requirements
  - Mass, cost, etc.
  - Intrusion of the battery tray during critical load case of side pole impact
  - These parameters will be used to design the future composite battery enclosure assembly





## Test matrix for material development

Variables	Factor 1	Factor 2	Factor3	Factor 4	<b>Total Factors</b>		
Layup	(0/45/-45/90/90/-45/45/0)						
CF/GF ratio	60%	45%	20%		3		
Thermoplastic Fiber	0	10%			2		
Co-mingling type	In-fiber tow	Lamina	Laminate-1	Laminate -2	4		
Stitch material	Thermoplastic				1		
Resin type	Epoxy				1		
Replicates					2		
Total combinations					48		

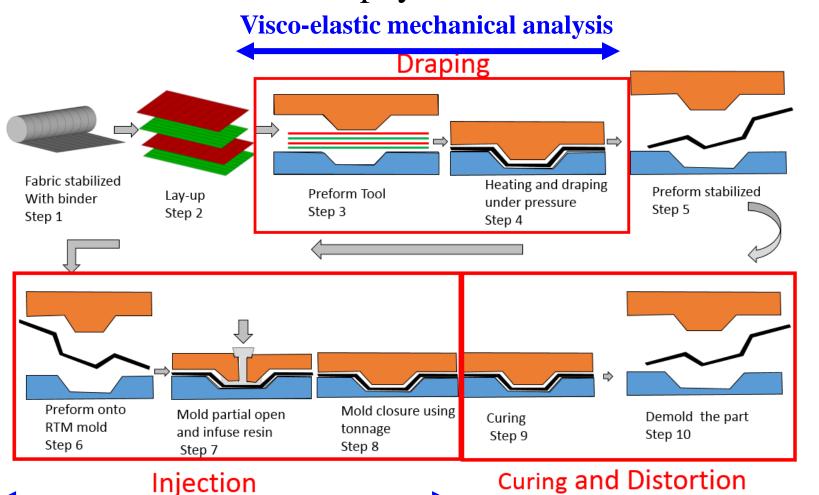
Lamina	5 CF tow bundles and 5 GF bundles side by side*	] ,
Laminate-1	$(0/45/-45/90)_s$ - 0, 90 layers are CF, and 45 and -45 layers are GF	
Laminate-2	(0/45/-45/90) <sub>s</sub> - 0, 90 layers are GF, and 45 and -45 layers are CF	]

\*Specimen width = 25 mm

## Manufacturing Process & Models

### HP-RTM Process - Multi-physics



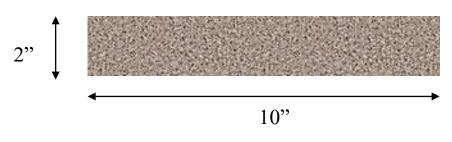


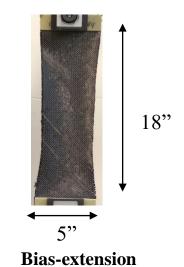
- Draping/Injection/ Curing stages
- Different physical processes need to be modeled for each manufacturing process.
- Output from one process is an input to the next process.

Fluid flow in porous media, heat Chemical kinetics followed by transfer, chemical kinetics – coupled mechanical analysis

### Testing matrix for draping model development







**Bending characterization** 

Dry fabrics needed to test for both the bending and bias-extension experiments

Variables	Factor 1	Factor 2	Factor3	Factor 4	Total Factors		
Co-mingling	Tow	Lamina	amina Laminate		3		
Layup		1					
	Bending tes	Bending tests - 0/90/90/0 and 90/0/0/90 with 0 deg. along the 10" side					
Material		1					
Spacing of fiber bundles	Standard	Higher			2		
Stitch density	]	Low High			2		
Stitch material		2					
Base material	50 gsm. GF		TP		1		
<b>Total combinations</b>					48		

## **Testing matrix for injection model**



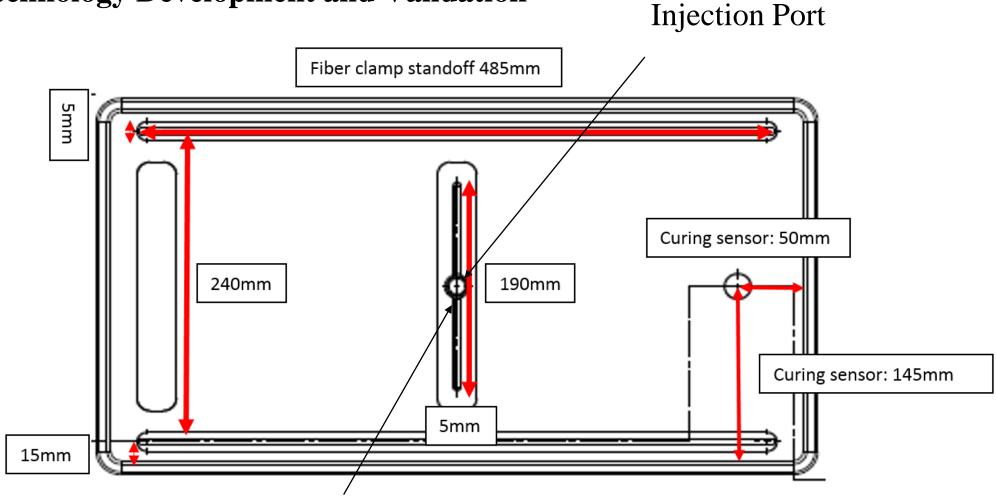


Permeability measuring instrument

Variables	Factor 1	Factor 2	Factor3	Factor 4	Total Factors
Layup	0	45	90		3
CF/GF volume ratio	60%	45%	20%		3
Spacing of fiber bundles	Standard		2		
Stitch density	Low		2		
Base material		50	gsm. GF		1
Replicates	1				1
<b>Total combinations</b>					36



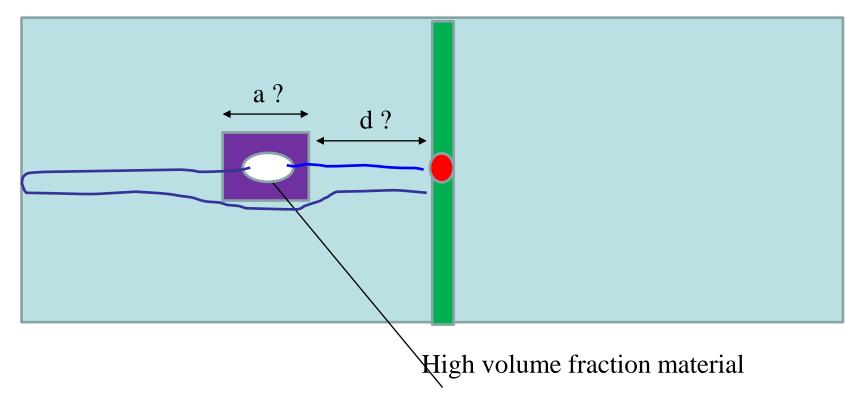
AI/ML Technology Development and Validation



Pressure sensor

#### AI/ML Technology Development and Validation

Create an intentional dry spot and use AI/ML technology to mitigate/remedy the defect (modifying the process conditions on the fly)



Experimental setup for AI/ML demonstration

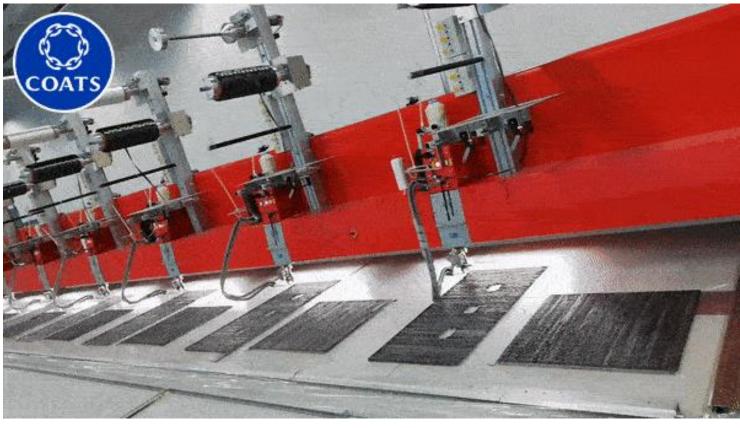


#### Tasks:

- dimension, and the volume fraction of the obstruction to create a void from simulations
- 2) Mold the samples to test the condition
- 3) Develop the AI/ML models that can predict the process conditions to mitigate/eliminate the dry spot following an indication on the pressure sensor reading
- 4) Validate the developed AI/ML in a demonstration exercise

## Unique Facilities











## **GM HP-RTM System**

- 1000T press
- Commercial injection system



## Responses to Previous Year Reviewers' Comments

• As the project was in the first year, no reviews were made.

## Partners/Collaborators



## General Motors

- Lead PM
- Baseline Steel assembly
- Design of composite assembly
- HP-RTM reinforcement
- Structural Design
- Structural testing/ validation•

#### Coats

 Tier 2, technology leader in TFP

#### **ESI**

Virtual prototype software development company, Global technology leader

## Columbia University

 Strong expertise in sensors, energy harvesting technology and data analytics

# Michigan State University (IACMI)

• State-of-the-art federally funded facilities for composite manufacturing

# Continental Structural Plastics

- Key Tier1 supplier
- Process design
- FR material development

# University of Southern California

- Expert at AI/ML
- DOE SciDAC institute

## Remaining Challenges and Barriers



(Any proposed future work is subject to change based on funding levels)

- Finalize structural design of the composite battery enclosure
- Finalize manufacturing process design of the composite assembly
- Finalize AI/ML technology for monitoring the manufacturing process of composite assembly
- Build the manufacturing tools
- Initial manufacturing of components of the assembly

## Summary

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- Due to Covid-19 and contractual delays, the project could only be started beginning April 1, 2021.
- Functional requirements for the baseline battery enclosure were determined.
- A full factorial test matrix was developed for optimizing the proposed hybrid fiber material systems and develop the structural performance model. Composite preforms are being manufactured with the above permutation and combination.
- Characterization plan for the conductive wiring material was completed for the self-health monitoring technology.
- An experimental setup was designed for the development and validation of the AI/ML process monitoring technology.



## Thank You!

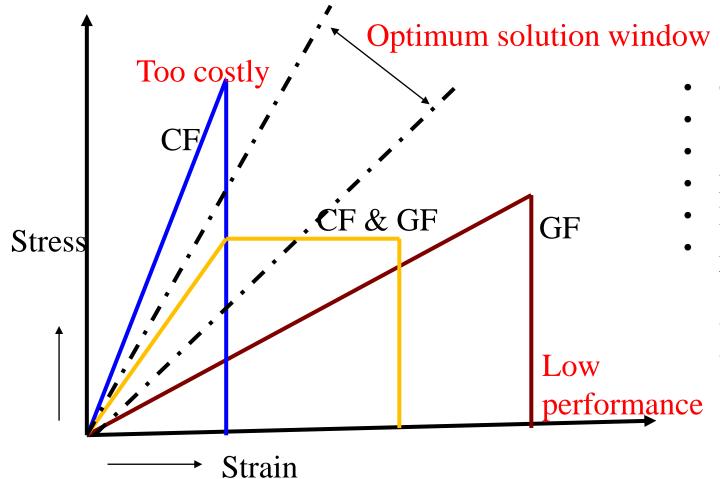


## Technical Back-Up Slides

## Hybrid Fiber Composites

## **Tension**





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- Strength ratio of GF\*/CF\$ = 3448/4157 = 0.83
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- Load case interest strength based
- Domain of interest beyond the initial failure
- A flexure, crashworthiness performance case will be interesting with more glass fiber content as glass fibers when compared to carbon fibers has good strength properties compared to stiffness.

## Governing Equations in Injection, Curing and Warpage



#### Filling – Stage – Coupled flow, heat and cure

Darcy's equation – Fluid Flow 
$$\nabla \cdot (-\frac{K}{\mu} \overrightarrow{\nabla P}) = 0$$

$$\text{Heat Transfer Equation} \quad \rho \mathcal{C}_p \tfrac{\partial T}{\partial t} + \, \rho_r \mathcal{C}_{pr} V \cdot \nabla T = \nabla \cdot (k \cdot \nabla T) - \rho_r \, \Delta h \, \tfrac{d\alpha}{dt}$$

Curing Kinetics 
$$\frac{d\alpha}{dt} = \left(A_1 \exp\left(-\frac{E_1}{T}\right) + A_2 \exp\left(-\frac{E_2}{T}\right) \cdot \alpha^m\right) \cdot \left(B - \alpha\right)^n$$

#### Curing - Stage - Coupled heat and cure

Heat Transfer Equation 
$$\rho C_p \frac{\partial T}{\partial t} + \rho_r C_{pr} V \cdot \nabla T = \nabla \cdot (k \cdot \nabla T) - \rho_r \Delta h \frac{d\alpha}{dt}$$
Curing Kinetics 
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#### Distortion- Stage (Thermo- Chemical Mechanical Analysis)

$$\sigma_{ij}(t) = \int_0^t C_{ijkl}(\xi(t) - \xi(\tau)) \frac{\partial \left(\epsilon_{kl} - \epsilon_{kl}^E\right)}{\partial \tau} d\tau \qquad C_{ijkl}(t) = \begin{cases} 0 & , X < X_{gel} \\ C_{ijkl}^{\infty} + \sum_{p=1}^P C_{ijkl}^p \cdot \left(e^{-t/\rho_{ijkl}^P}\right), X \ge X_{gel} \end{cases}, \text{no sum on } i, j, k, l$$

Di Benedetto function  $\rightarrow T_g$ 

$$\frac{T_g - T_{g0}}{T_{g\infty} - T_{g0}} = \frac{\lambda X}{1 - (1 - \lambda)X}$$

